import matplotlib.pyplot as plt In [2]: df=pd.read_csv('Social_Network_Ads.csv') Out[2]: User ID Gender Age EstimatedSalary Purchased **0** 15624510 19000 Male 19 **1** 15810944 Male 35 20000 0 **2** 15668575 Female 26 43000 0 57000 **3** 15603246 Female 27 **4** 15804002 Male 19 76000 0 **395** 15691863 Female 46 41000 1 **396** 15706071 51 23000 Male 1 20000 **397** 15654296 Female 1 33000 **398** 15755018 Male 36 0 **399** 15594041 Female 36000 1 400 rows × 5 columns **Explore Dataset** In [3]: df.head() User ID Gender Age EstimatedSalary Purchased Out[3]: **0** 15624510 Male 19 19000 0 **1** 15810944 Male 35 20000 0 **2** 15668575 Female 0 26 43000 **3** 15603246 Female 27 57000 **4** 15804002 76000 0 19 Male In [4]: df.tail() User ID Gender Age EstimatedSalary Purchased Out[4]: **395** 15691863 Female 41000 1 46 **396** 15706071 Male 51 23000 **397** 15654296 Female 50 20000 1 **398** 15755018 33000 Male 0 **399** 15594041 Female 36000 1 df.describe <bound method NDFrame.describe of</pre> User ID Gender Age EstimatedSalary Purchased 15624510 Male 19000 0 19 15810944 35 20000 0 Male 15668575 Female 26 43000 0 15603246 Female 27 57000 0 15804002 19 76000 . . . 395 15691863 Female 41000 15706071 396 23000 397 15654296 Female 50 20000 398 15755018 Male 36 33000 0 15594041 Female 49 36000 1 [400 rows x 5 columns]> In [6]: df.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 400 entries, 0 to 399 Data columns (total 5 columns): Column Non-Null Count Dtype User ID 0 400 non-null int64 1 Gender 400 non-null object 2 400 non-null int64 EstimatedSalary 400 non-null int64 3 400 non-null int64 Purchased dtypes: int64(4), object(1) memory usage: 15.8+ KB df.dtypes int64 User ID Gender object Age int64 EstimatedSalary int64 Purchased int64 dtype: object In [8]: df.isnull().sum() User ID Out[8]: Gender 0 Age EstimatedSalary 0 Purchased dtype: int64 df.duplicated().sum() Out[9]: In [10]: mapi={'Male':1, 'Female':0} df=df.replace(mapi) df.head() Out[10]: User ID Gender Age EstimatedSalary Purchased **0** 15624510 1 19 19000 0 1 35 **1** 15810944 20000 0 43000 0 **2** 15668575 0 26 **3** 15603246 0 27 57000 0 **4** 15804002 76000 0 1 19 In [11]: df.drop(['User ID'],axis=1, inplace=True) df.head() Out[11]: Gender Age EstimatedSalary Purchased 1 19 0 19000 2 43000 0 0 26 57000 0 0 1 19 76000 Train test split In [12]: x, y=df.drop(['Purchased'], axis=1), df['Purchased'] from sklearn.model_selection import train_test_split xtrain, xtest, ytrain, ytest=train_test_split(x, y, test_size=0.25, random_state=0) Standard scaler In [14]: from sklearn.preprocessing import StandardScaler sc_scale=StandardScaler() xtrain=sc_scale.fit_transform(xtrain) xtest=sc_scale.transform(xtest) In [15]: from sklearn.naive_bayes import GaussianNB classifier=GaussianNB() classifier.fit(xtrain,ytrain) Out[15]: ▼ GaussianNB GaussianNB() In [16]: y_pred=classifier.predict(xtest) confusion matrix In [17]: from sklearn.metrics import confusion_matrix cm=confusion_matrix(ytest,y_pred) print("Confusion matrix is : \n",cm) Confusion matrix is : [[66 2] [7 25]] In [18]: import seaborn as sns import matplotlib.pyplot as plt sns.heatmap(cm, annot=True) plt.show() - 60 - 30 Accuracy and score In [19]: from sklearn.metrics import accuracy_score print("Accuracy :", accuracy_score(ytest,y_pred)*100,'%') Accuracy : 91.0 % In [20]: from sklearn.metrics import precision_score from sklearn.metrics import recall_score from sklearn.metrics import f1_score In [22]: #precision tp/(tp+fp) precision=precision_score(ytest,y_pred) print('Precision: %f' % precision) #recall tp/(tp+fn) recall_recall_score(ytest,y_pred) print('recall: %f' % recall) #precision 2tp/(2tp+fp+fn) f1=f1_score(ytest,y_pred) print('f1: %f' % f1)

import pandas as pd
import numpy as np

Precision: 0.925926 recall: 0.781250 f1: 0.847458